

$$a. \quad -\sum_{w \in \text{Vocab}} y_w \log(\hat{y}_w) = -(y_1 \log(\hat{y}_1) + \dots + y_0 \log(\hat{y}_0) + \dots + y_{|V|} \log(\hat{y}_{|V|}))$$

$$\therefore y_w = \begin{cases} 1 & w=0 \\ 0 & w \neq 0 \end{cases}$$

$$\therefore \text{原式} = -y_0 \log(\hat{y}_0) = -\log(\hat{y}_0)$$

$$b. \quad \frac{\partial J_{\text{naive-softmax}}(V_c; \mathbf{0}, U)}{\partial V_c} = \frac{\partial}{\partial V_c} \left(-\log \frac{\exp(\mathbf{U}_0^T V_c)}{\sum_{w \in \text{Vocab}} \exp(\mathbf{U}_w^T V_c)} \right)$$

$$\frac{\partial}{\partial V_c} \text{原式} = \frac{\partial}{\partial V_c} (-\log \exp(\mathbf{U}_0^T V_c) + \log \sum_{w \in \text{Vocab}} \exp(\mathbf{U}_w^T V_c))$$

$$= \frac{\partial}{\partial V_c} (-\mathbf{U}_0^T V_c) + \log \sum_{w \in \text{Vocab}} \exp(\mathbf{U}_w^T V_c)$$

$$= -\mathbf{U}_0 + \frac{\frac{\partial}{\partial V_c} (\sum_{w \in \text{Vocab}} \exp(\mathbf{U}_w^T V_c))}{\sum_{w \in \text{Vocab}} \exp(\mathbf{U}_w^T V_c)}$$

$$= -\mathbf{U}_0 + \frac{\sum_{w \in \text{Vocab}} [\exp(\mathbf{U}_w^T V_c) \cdot \mathbf{U}_w]}{\sum_{w \in \text{Vocab}} \exp(\mathbf{U}_w^T V_c)}$$

$$= -\mathbf{U}_0 + \sum_{w \in \text{Vocab}} \hat{y}_w \mathbf{U}_w$$

$$= \sum_{w \in \text{Vocab}} (\hat{y}_w \mathbf{U}_w - y_w \mathbf{U}_w)$$

$$= \sum_{w \in \text{Vocab}} (\hat{y}_w - y_w) \mathbf{U}_w = U(\hat{y} - y)$$

$$c. \quad \frac{\partial J_{\text{naive-softmax}}(V_c, \mathbf{0}, U)}{\partial U_w} = \frac{\partial (-\log \exp(\mathbf{U}_0^T V_c) + \log \sum_{w \in \text{Vocab}} \exp(\mathbf{U}_w^T V_c))}{\partial U_w}$$

$$= \frac{\partial \mathbf{U}_0^T}{\partial U_w} \cdot V_c + \frac{\sum_{w \in \text{Vocab}} (\exp(\mathbf{U}_w^T V_c) \cdot V_c)}{\sum_{w \in \text{Vocab}} \exp(\mathbf{U}_w^T V_c)}$$

$$\frac{\partial \mathbf{U}_0^T}{\partial U_w} = \begin{cases} 1 & w=0 \\ 0 & w \neq 0 \end{cases} \quad \therefore \frac{\partial \mathbf{U}_0^T}{\partial U_w} = y_w = \sum_{w \in \text{Vocab}} y_w$$

$$\therefore \text{原式} = \sum_{w \in \text{Vocab}} \left[-y_w V_c + \frac{\exp(\mathbf{U}_w^T V_c)}{\sum_{w \in \text{Vocab}} \exp(\mathbf{U}_w^T V_c)} V_c \right]$$

$$= \sum_{w \in \text{Vocab}} V_c (\hat{y}_w - y_w)$$

$$= \begin{cases} V_c (\hat{y}_0 - 1) & w=0 \\ V_c \hat{y}_w & w \neq 0 \end{cases}$$

$$d. \quad \partial J_{\text{naive-sfomax}}(v_c, o, u) = \left[\frac{\partial J}{\partial u_1}, \frac{\partial J}{\partial u_2}, \dots, \frac{\partial J}{\partial u_M} \right]$$

$$e. \quad \frac{\partial \sigma}{\partial x} = \frac{e^x(e^x+1) - e^x e^x}{(e^x+1)^2} = \frac{e^x}{(e^x+1)^2} = \sigma(x) \cdot (1-\sigma(x))$$

$$f. \quad (1) \quad \frac{\partial J_{\text{neg-sample}}(v_c, o, u)}{\partial v_c} = \frac{\partial}{\partial v_c} \left[-\log(\sigma(u_0^T v_c)) - \sum_{k=1}^K \log(\sigma(-u_k^T v_c)) \right]$$

$$= \left(-\frac{1}{\sigma(u_0^T v_c)} \right) (u_0 \sigma(u_0^T v_c) (1 - \sigma(u_0^T v_c)))$$

$$- \sum_{k=1}^K \frac{-u_k \sigma(-u_k^T v_c) (1 - \sigma(-u_k^T v_c))}{\sigma(-u_k^T v_c)}$$

$$= -u_0 (1 - \sigma(u_0^T v_c)) + \sum_{k=1}^K u_k (\sigma(-u_k^T v_c))$$

$$= u_0 (\sigma(u_0^T v_c) - 1) - \sum_{k=1}^K u_k (\sigma(-u_k^T v_c) - 1)$$

$$(2) \quad \frac{\partial J_{\text{neg-sample}}(v_c, o, u)}{\partial u_0} = \frac{\partial}{\partial u_0} \left[-\log(\sigma(u_0^T v_c)) - \sum_{k=1}^K \log(\sigma(-u_k^T v_c)) \right]$$

$$= \frac{\partial}{\partial u_0} [-\log(\sigma(u_0^T v_c))] - \frac{\partial}{\partial u_0} \left[\sum_{k=1}^K \log(\sigma(-u_k^T v_c)) \right]$$

$$= \frac{\partial}{\partial u_0} [-\log(\sigma(u_0^T v_c))]$$

$$= -\frac{1}{\sigma(u_0^T v_c)} (\sigma(u_0^T v_c) (1 - \sigma(u_0^T v_c)) v_c) \neq 0$$

$$= v_c (\sigma(u_0^T v_c) H)$$

$$(3) \quad \frac{\partial J_{\text{neg-sample}}(v_c, o, u)}{\partial u_k} = 0 - \frac{\partial}{\partial u_k} \left[\sum_{k=1}^K \log(\sigma(-u_k^T v_c)) \right]$$

$$= \frac{-v_c \sigma(-u_k^T v_c) (1 - \sigma(-u_k^T v_c))}{\sigma(-u_k^T v_c)} = v_c (1 - \sigma(-u_k^T v_c))$$

g. $\frac{\partial \text{Neg-sample}(V_c; 0, U)}{\partial u_k} = \frac{\partial}{\partial u_k} \left[-\log(\sigma(u_0^T V_c)) - \sum_{k=1}^K \log(\sigma(-u_k^T V_c)) \right]$

$$= \frac{\partial}{\partial u_k} [-\log(\sigma(u_0^T V_c))] - \frac{\partial}{\partial u_k} \left[\sum_{k=1}^K \log(\sigma(-u_k^T V_c)) \right]$$

$$= 0 - \frac{\partial}{\partial u_k} \left[\sum_{k=1}^K \log(\sigma(-u_k^T V_c)) \right]$$

$$= - \left[\sum_{i \in \{1, \dots, K\}: w_i = w_k} \frac{\partial}{\partial u_k} \log(\sigma(-u_i^T V_c)) + \sum_{i \in \{1, \dots, K\}: w_i \neq w_k} \frac{\partial}{\partial u_k} \log(\sigma(-u_{i, w_i}^T V_c)) \right]$$

$$= - \sum_{i \in \{1, \dots, K\}: w_i = w_k} \frac{\partial}{\partial u_k} \log(\sigma(-u_i^T V_c))$$

$$= - \sum_{i \in \{1, \dots, K\}: w_i = w_k} V_c (1 - \sigma(-u_k^T V_c))$$

h. (1) $\frac{\partial \text{Skip-gram}(V_c, w_{t-m}, \dots, w_{t+m}, U)}{\partial u} = \sum_{\substack{-m \leq j \leq m \\ j \neq 0}} \frac{\partial J(V_c, w_{t+j}, U)}{\partial u}$

(2) $\frac{\partial \text{Skip-gram}(V_c, w_{t-m}, \dots, w_{t+m}, U)}{\partial V_c} = \sum_{\substack{-m \leq j \leq m \\ j \neq 0}} \frac{\partial J(V_c, w_{t+j}, U)}{\partial V_c}$

(3) $\frac{\partial \text{Skip-gram}(V_c, w_{t-m}, \dots, w_{t+m}, U)}{\partial V_w w_{t-c}} = 0$

